A Framework for UAV Sensor Fusion

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VAV – Unmaned Aerial Vehicle

No human supervision during long time spans

Autonomous

know their own state, operate to achieve goals





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Sensor fusion

- Sensors: imperfect observations about a system
- **Fusion:** combine information from different sources, so that the final result is better than what we could achieve using each source separately



Sensor fusion

Include knowledge about observed system to improve results



Sensor fusion

- In our case, this knowledge is a description of the dynamics of a plane
- Typically, evolution modeled as a first order Markov process
- Uncertainty sources:
 - Model inaccuracies
 - Sensor measures are not perfect

Why a framework? - Motivation

Typical experimental methodology:



- Repeat steps 3–5 in a loopy fashion
 - Sometimes, new data is needed.

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Why a framework? - Motivation

Save work

- Implemented parts (structure, utilities)
- Make collaborative efforts possible
 - Architecture, modularity, data specification



Framework features

Centralized fusion

Single vehicle equipped with sensors

Passive, event-driven

- Do "computing stuff" on new events
- Periodical sensor updates, not under request

Implementation in MATLAB

- Fast prototyping (for research)
- Many useful tools

Architecture



Architecture



- Why simulation?
 - Independent of external conditions
 - Not unexpected technical issues
 - Fast evaluation of changes
 - Study "extreme" cases (crashes)





- Basic implementation:
 - High level specification of trajectory
 - Simple Six Degrees of Freedom (6DoF) system
 - Weight, inertia
 - Punctual mass in vacuum (no complex aerodynamics)
- Returns data about the UAV trajectory
 - Required by the sensor simulation module



Basic implementation:

Input to simulation module – high level specification of trajectory

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. . .

straight flight linear acceleration turn straight flight random turns (speed, duration)
(acceleration, duration)
(axis, angle, duration)
(speed, duration)
(number, angle limit, ...)



- Basic implementation:
 - Intermediate product: forces and moments
 - Array of tuples [time, values X,Y,Z]

Forces:						
time	force X	forceY	force Z			
00.00	3.24	-0.03	0.00			
02.40	-2.50	0.00	0.30			
31.50	0.00	0.00	0.00			
<u>Moments:</u>						
time	mmntX	mmntY	mmntZ			
05.00	0.00	0.00	0.00			
07.36	0.00	2.27	-1.45			
•••						



- Basic implementation:
 - Expected output: exact values for the physical attributes of interest
 - Fixed timestep, high frequency (adjustable)

Time	Position Acc	el. Speed	Ang. Rate
0.01	••••	•••	•••
0.02	••••	•••	•••
0.03	••••	•••	• • •
0.04	••••	•••	• • •
0.05	••• •••	•••	•••
••••			



Application: typical maneuvers in commercial flights



Can be easily extended/substituted

- FlightGear + JSBSim
 - Live trajectory generation
 - Realistic aerodynamics
 - Atmospheric conditions
 - Output to file



...and we can use real data!



- Input: data from simulation module
- **Output:** the measures that a real sensor would display
- Several devices already implemented
 - Matlab classes/functions
- Configurable parameters
 - "amount of error", update rate...

- Emulate inaccuracies and other effects of real life sensors
- Examples:
 - Noise: independent random perturbations of measures





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- Emulate inaccuracies and other effects of real life sensors
- Examples:
 - Noise: independent random perturbations of measures
 - Bias: persistent, almost-constant deviation between real value and measure
 - Correlated errors: physical effects that affect several components of a measure at the same time
 - **Quantization:** devices measuring a continuous attribute, but their output vary discretely

Example: barometric altimeter simulation



Architecture – sensor fusion



Architecture – sensor fusion

- Flexibility!
- Sample "template" scripts
- Implementation of classical techniques:
 - Kalman Filter (KF)
 - Extended KF
 - Unscented KF
 - Particle Filter
- "Soft interface":
 - Respect particularities of each technique
 - Minimize impact when switching amongst them



Numerical output

- Statistical validation
- Performance indices

Visual output

- System evolution <u>during</u> experiments
- Performance under special external conditions
- Comparison of algorithms/configurations

System evolution during experiments



Performance under special external conditions



Comparison of different algorithms/configurations



Conclusions

- Framework covers the whole experimental process
- Easy to try different parameters/algorithms
- Adding new parts is quite straightforward
- Focused on visual performance evaluation

Questions & suggestions

Thank you very much!